

“Graduated Embodiment for Sophisticated Agent Evolution and Optimization”

Position Paper on Scalability of Evolutionary Computation

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Graduated Embodiment for Sophisticated Agent Evolution and Optimization

Goal: *To develop methods for graduated evolution for embodied agent optimization.*

Method: *Evolve agent behaviors in a hierarchical fashion by staging fitness through multiple levels of fidelity.*



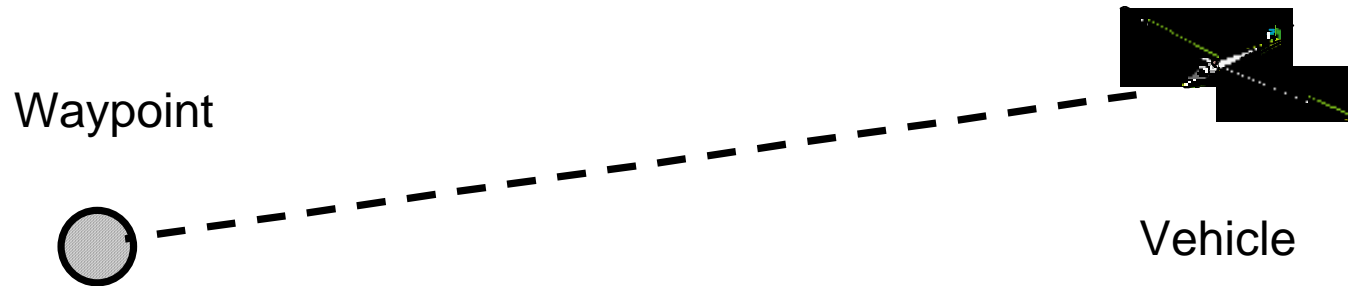
Project Accomplishments

- *Implemented Staged Optimization*
 - *Showed that staged methods exhibit improved scaling (as expected).*
- *Implemented Pruning*
 - *Pruned trees demonstrated reduced computational load (as expected).*
- *Combined hand-coded and machine-generated behaviors*
 - *Can build algorithms that are truly human-machine collaborations.*
- *Converted GP code to C++*
 - *Allows for improved modularization; shifts burden off user.*
- *Incorporated GP behaviors into UMBRA*
 - *High fidelity environment for high fidelity behaviors.*



Adaptive Waypoints

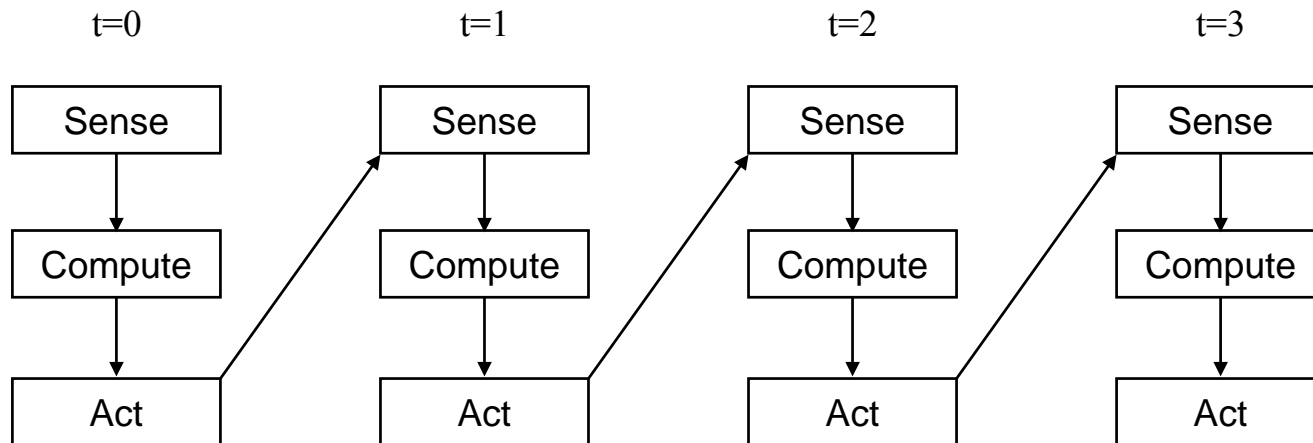
This encapsulation is intended to bridge the gap between toy problems and real problems



Our method to develop levels of integration:
Vertical, Horizontal, Co-Evolution, Collective Behavior

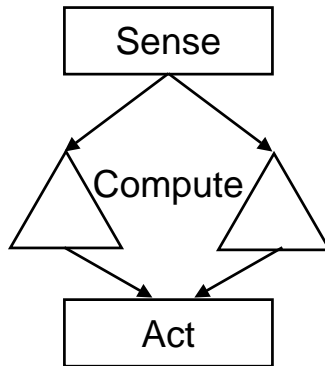


General implementation concept: Use GP to design “compute” step

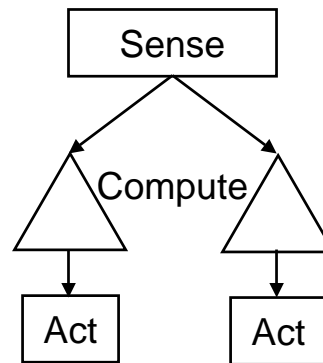


Classification of Building Blocks

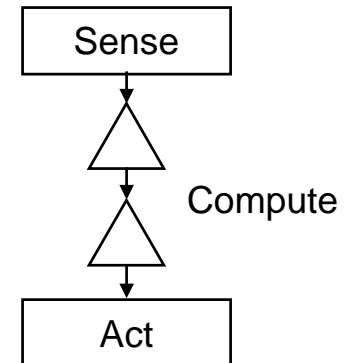
Horizontal Sequential



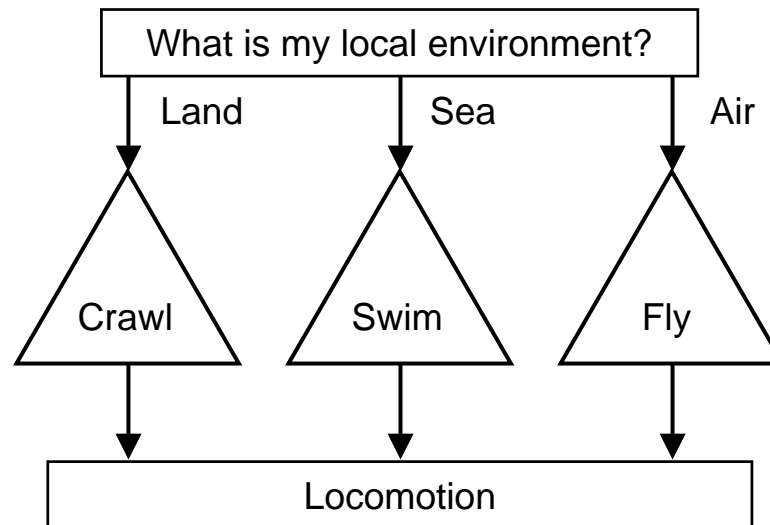
Horizontal Parallel



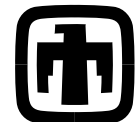
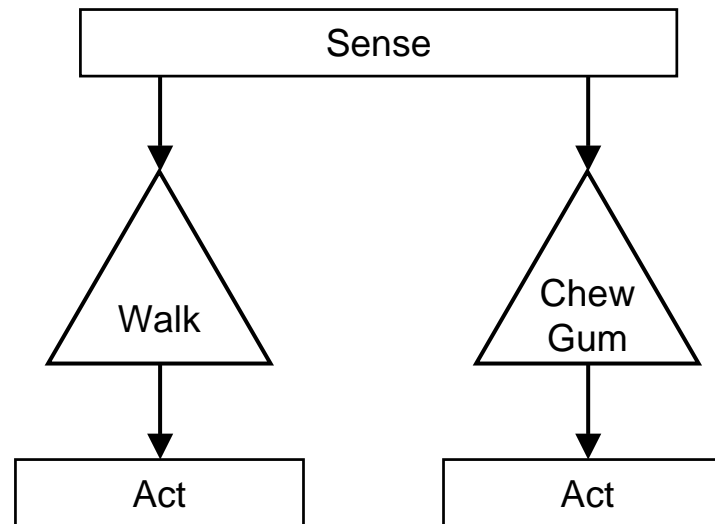
Vertical



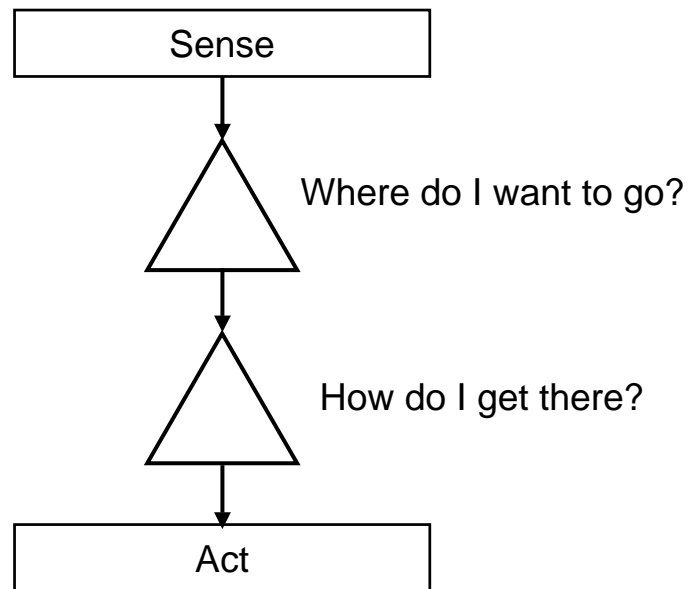
Example: Horizontal Sequential



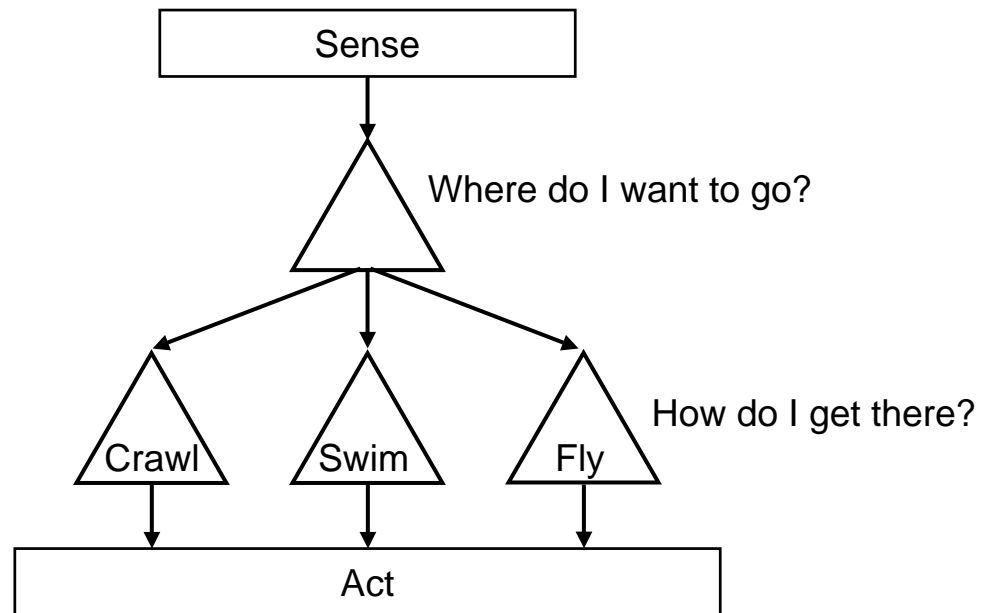
Example: Horizontal Parallel



Example: Vertical

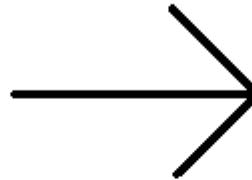


Example: Combined



Benefits of Converting code from C to C++: Modularization

C++ reduces and modularizes changes in code



Change optimization in C code

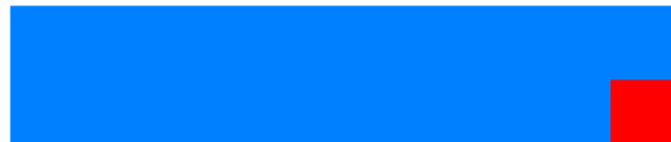
■ Code that stays the same

■ Code that changes

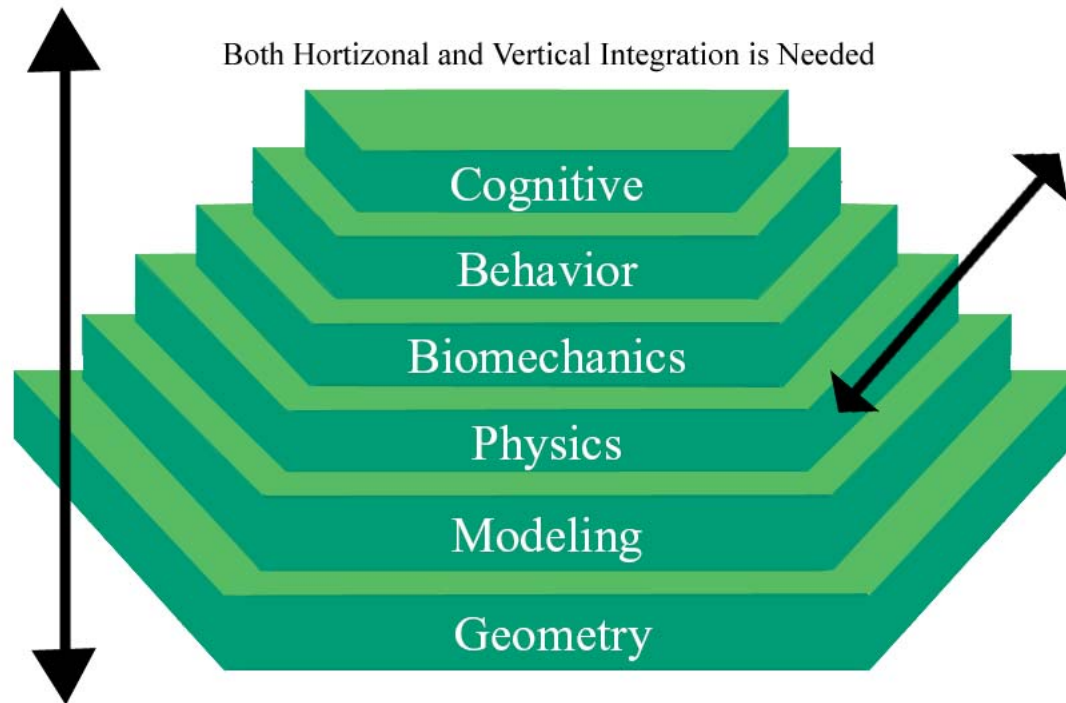


More changes in more places
mean more bugs and longer
development time

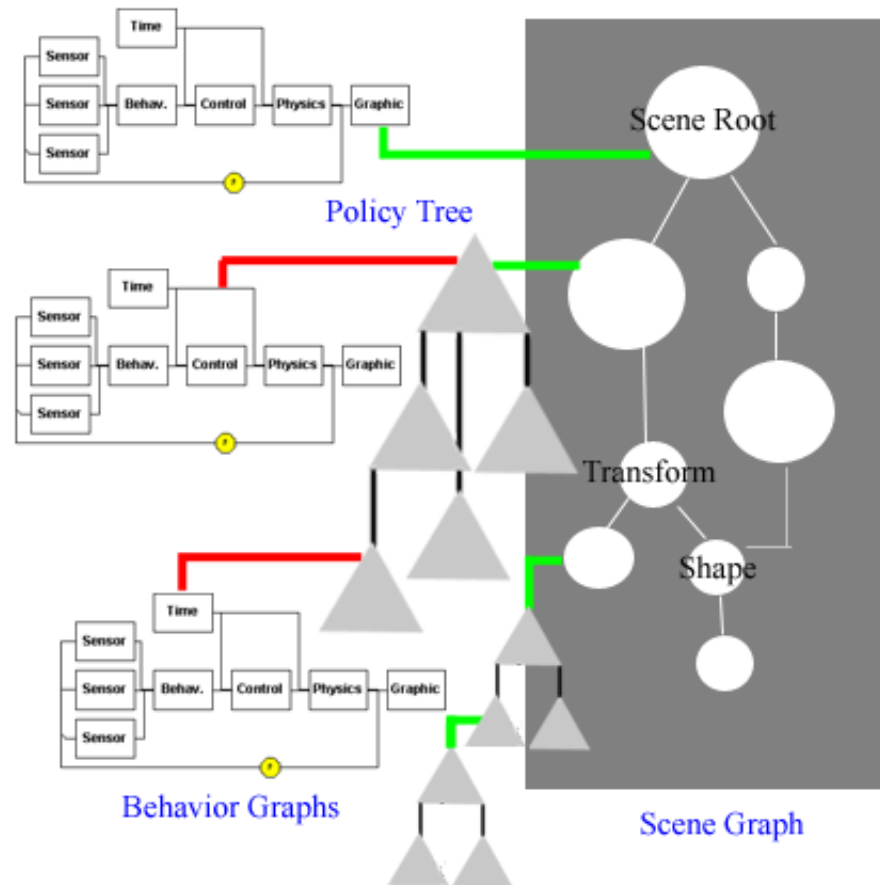
Change optimization in C++ code



Multiple levels are needed: both horizontal and vertical

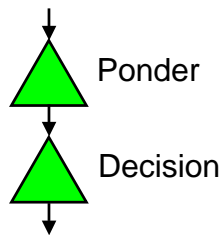


We are developing tools to visualize evolved behaviors

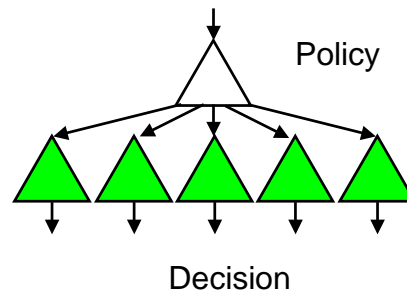


Benchmark problem optimization approaches

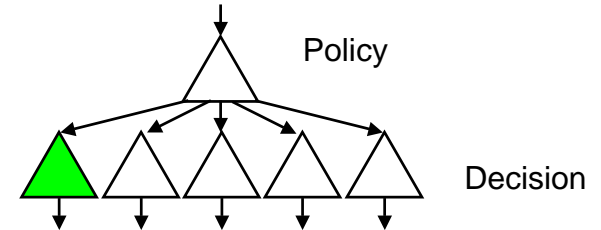
Single Tree Ponder



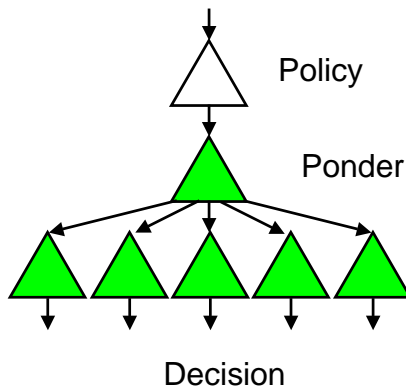
Five Tree No-Ponder



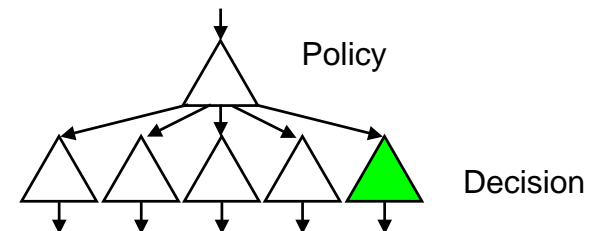
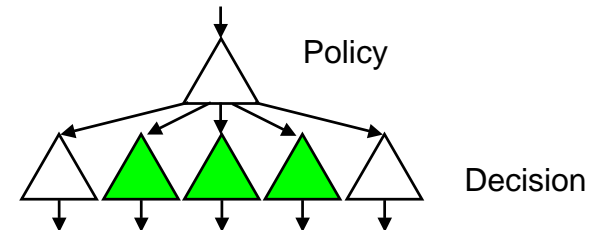
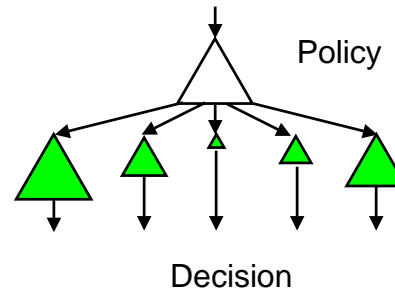
Staged Optimization



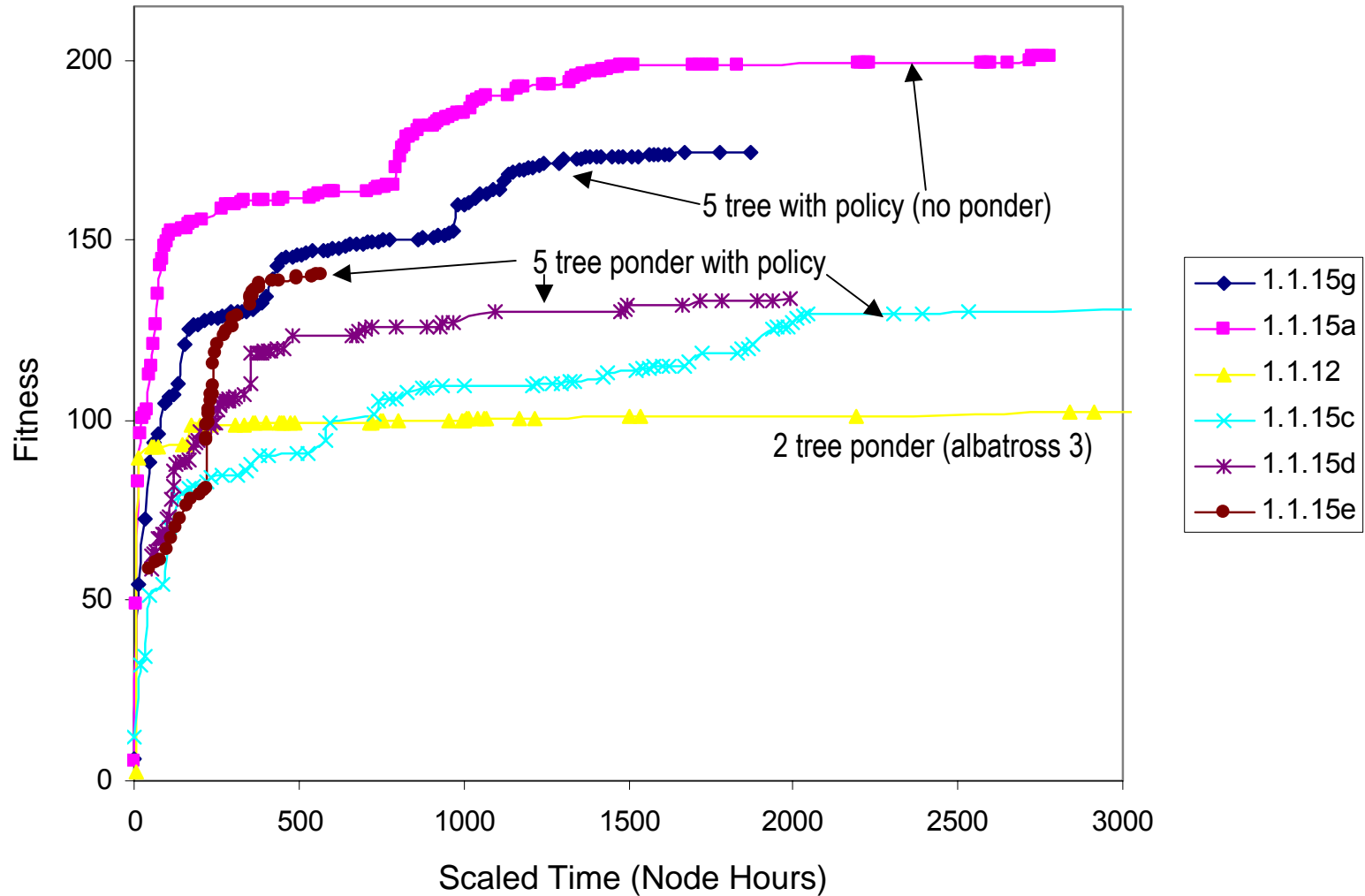
Five Tree Ponder



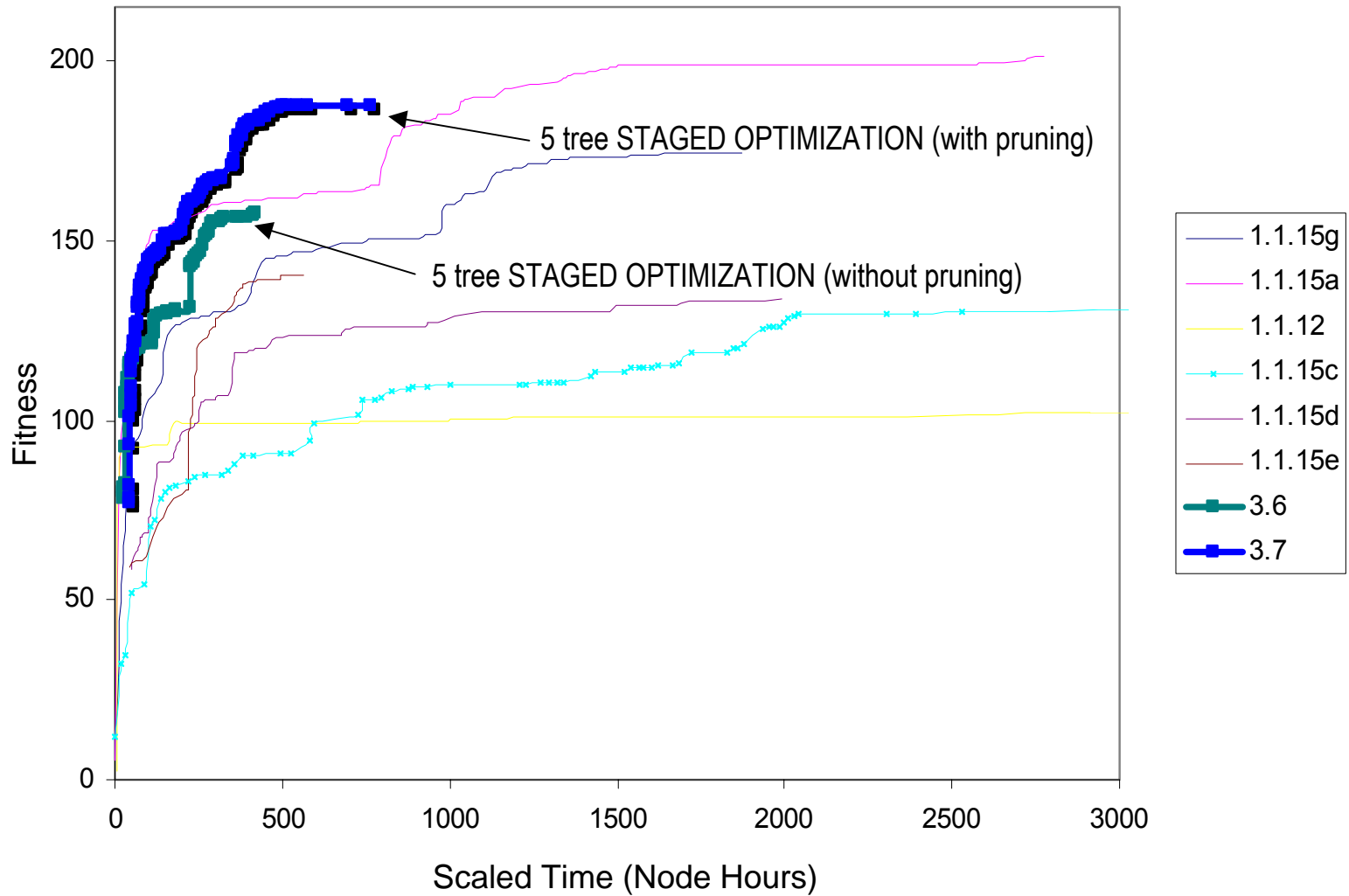
With Tree Pruning



Results for benchmark problem



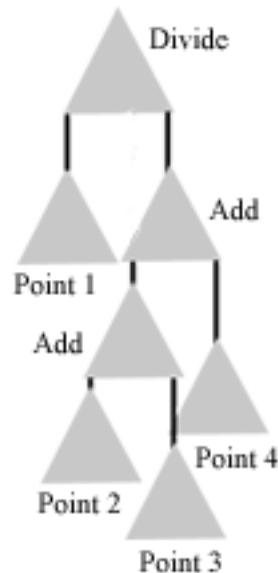
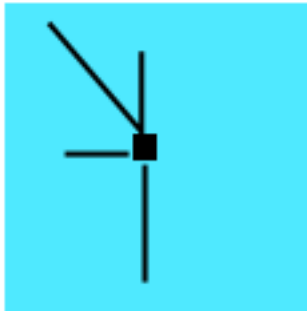
Results for benchmark problem



Real-world example: Image data collection

Sense Compute Act (Machines)

Pixel Neighborhood



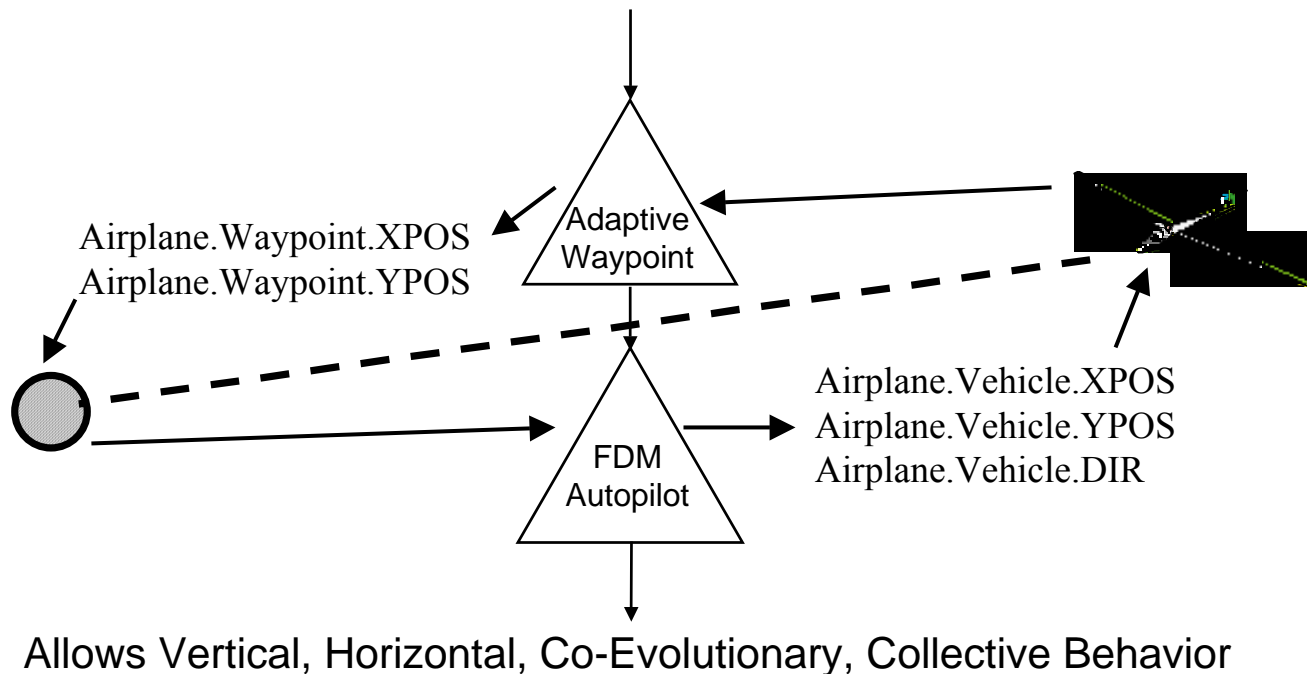
If Value < Target Value

MOVE (C)

Pereceptual Cognitive Behavioral (Humans)

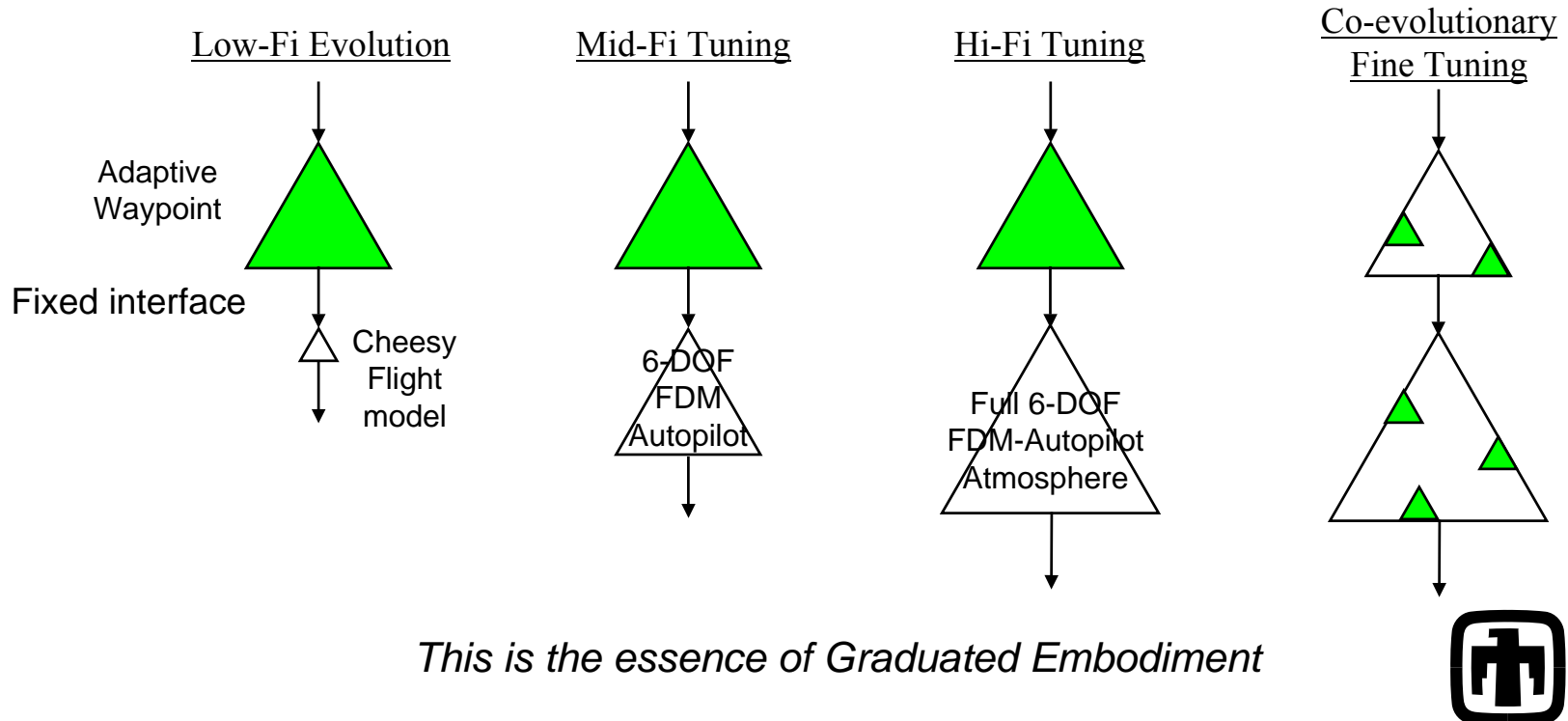


Graduated Embodiment example: adaptive waypoint



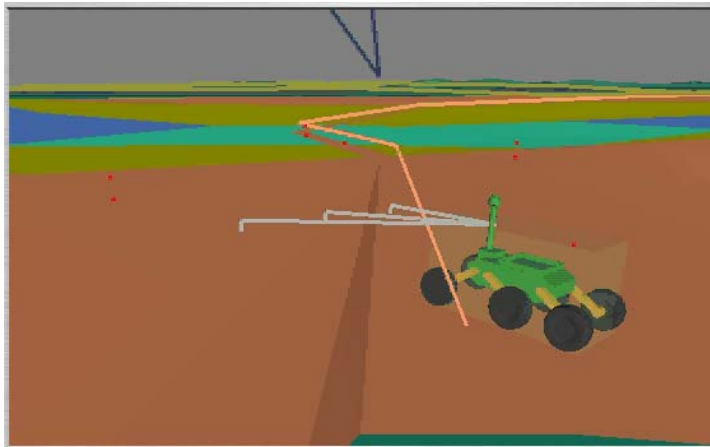
Graduated Embodiment example: adaptive waypoint

We have chosen this method to develop usable behaviors

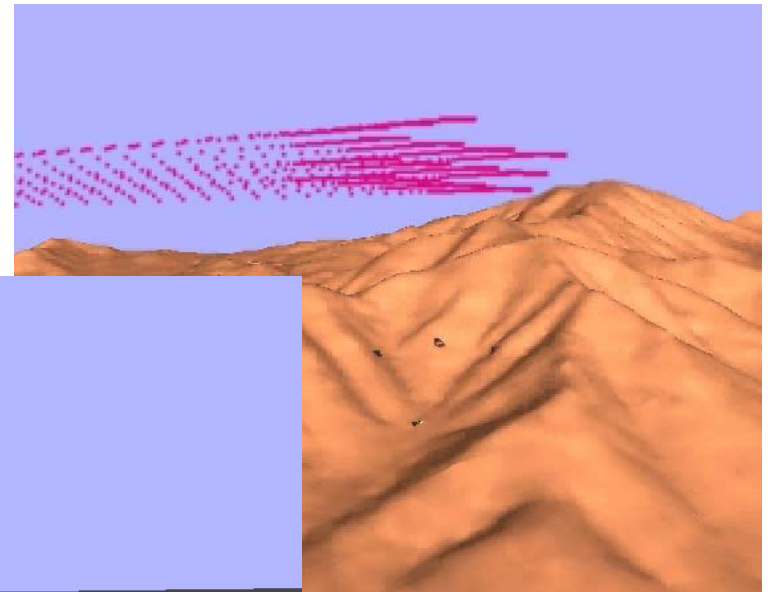


Next: Apply GP to Umbra's hard engineering

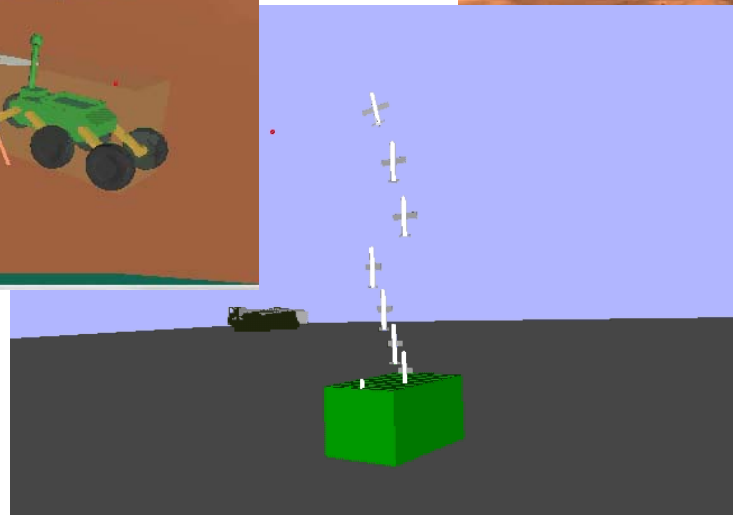
Can the adaptive waypoint concept be generalized to n-dimensional parameter space??



Simple target on terrain



Netfires target seeking



Netfires at launch

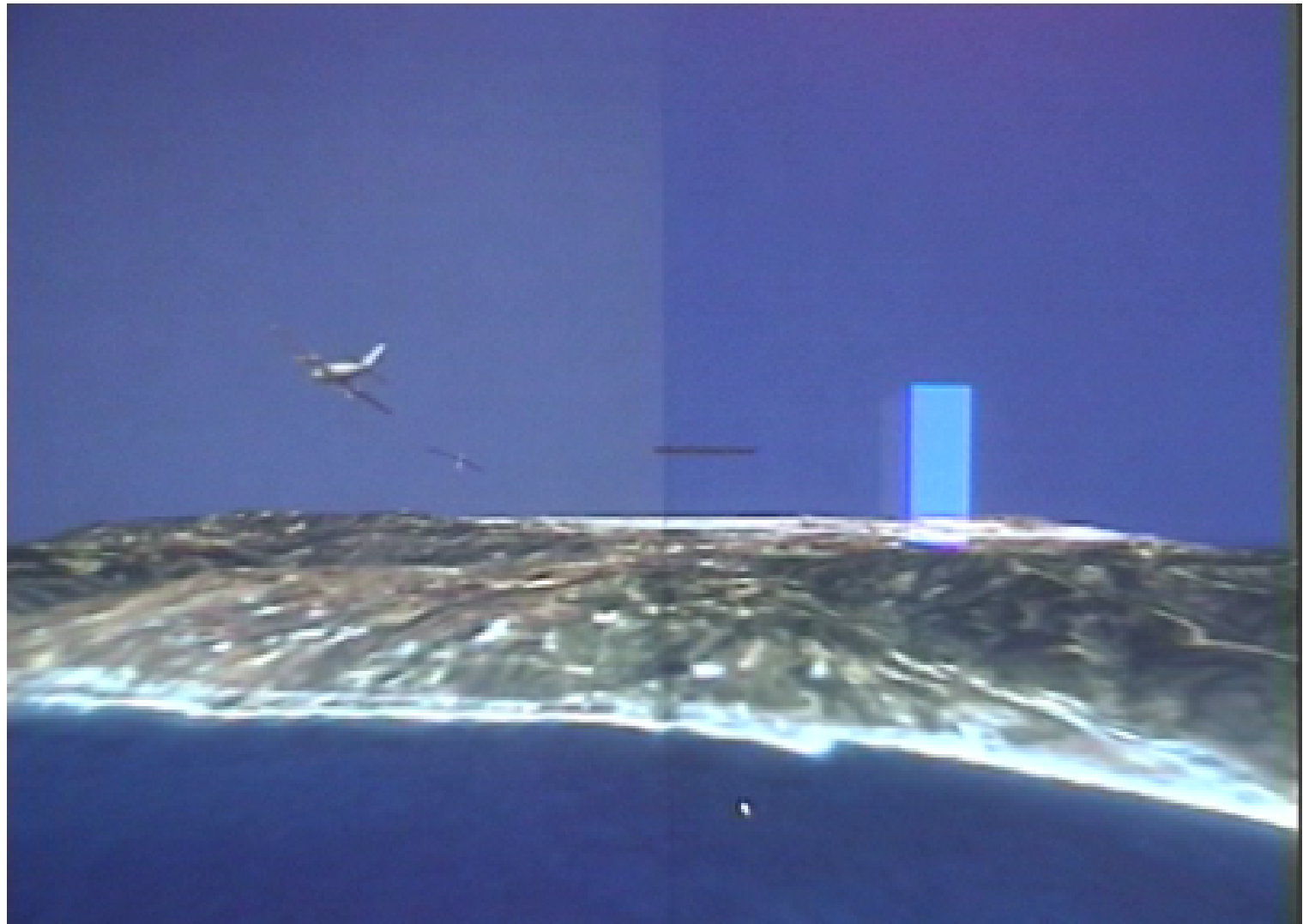


Why are we developing evolutionary computing technology?

- *Well-suited for robotics applications.*
- *Simulation pushes Umbra “multiple-fidelity” capability.*
- *Provides context for intelligent machine systems.*
- *Likely to provide insight into cognition processes.*
- *Uses behavioral biomimetics to derive insight from nature.*



Umbra Application of Genetic Programming Running on 2X2 Tile in VIEWS Corridor Building 880/A1



2056 X 2048 Resolution taken with 720X 480 Camera